

# **SOLAR IRRADIANCE NOWCASTING BASED ON A NETWORK OF ALL-SKY IMAGERS: THE VALUE OF HIGH-RESOLUTION DATA ON VARIABILITY INFORMATION**

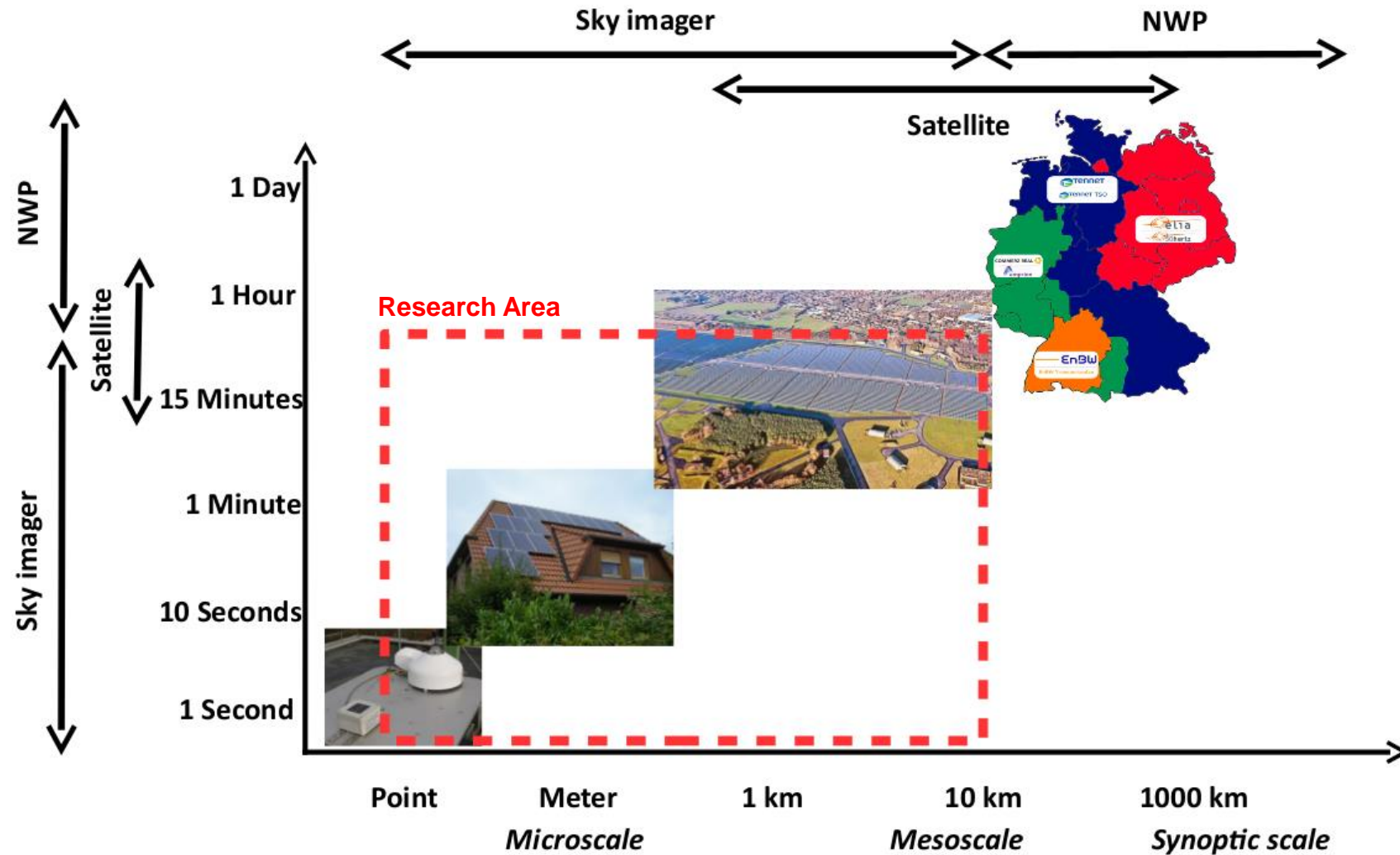
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**DLR Institute of Networked Energy Systems (\* and Institute of Solar Research)**



# Solar irradiance forecasts

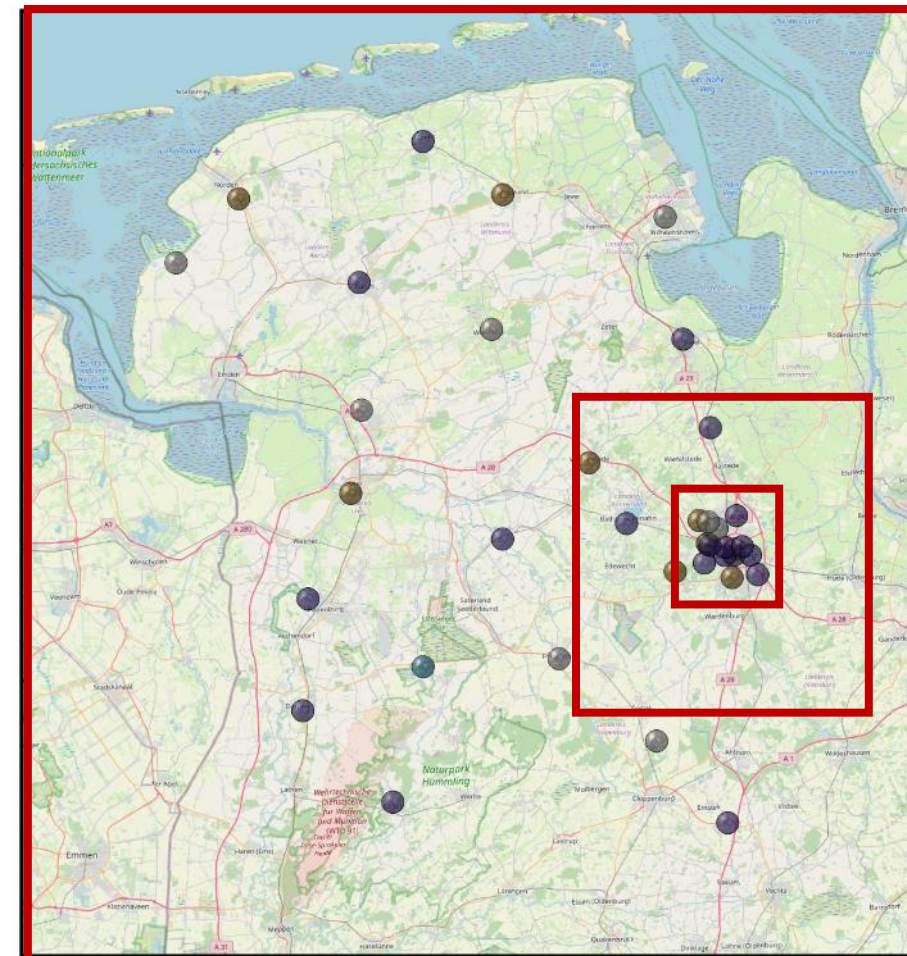
## Towards increasing spatial and temporal resolution



# Eye2Sky network

- 30 All-Sky Imager (ASI) installed in north-west Germany
  - With 12 stations equipped with meteorological equipment
- covering ~110km x 100km area in north-western Germany
- Low density in rural area covering low voltage distribution grid
- High station density in city of Oldenburg

## Eye2Sky - Cloud camera and meteorological measurement network in Oldenburg



# Instrumentation

## Meteorological sensors

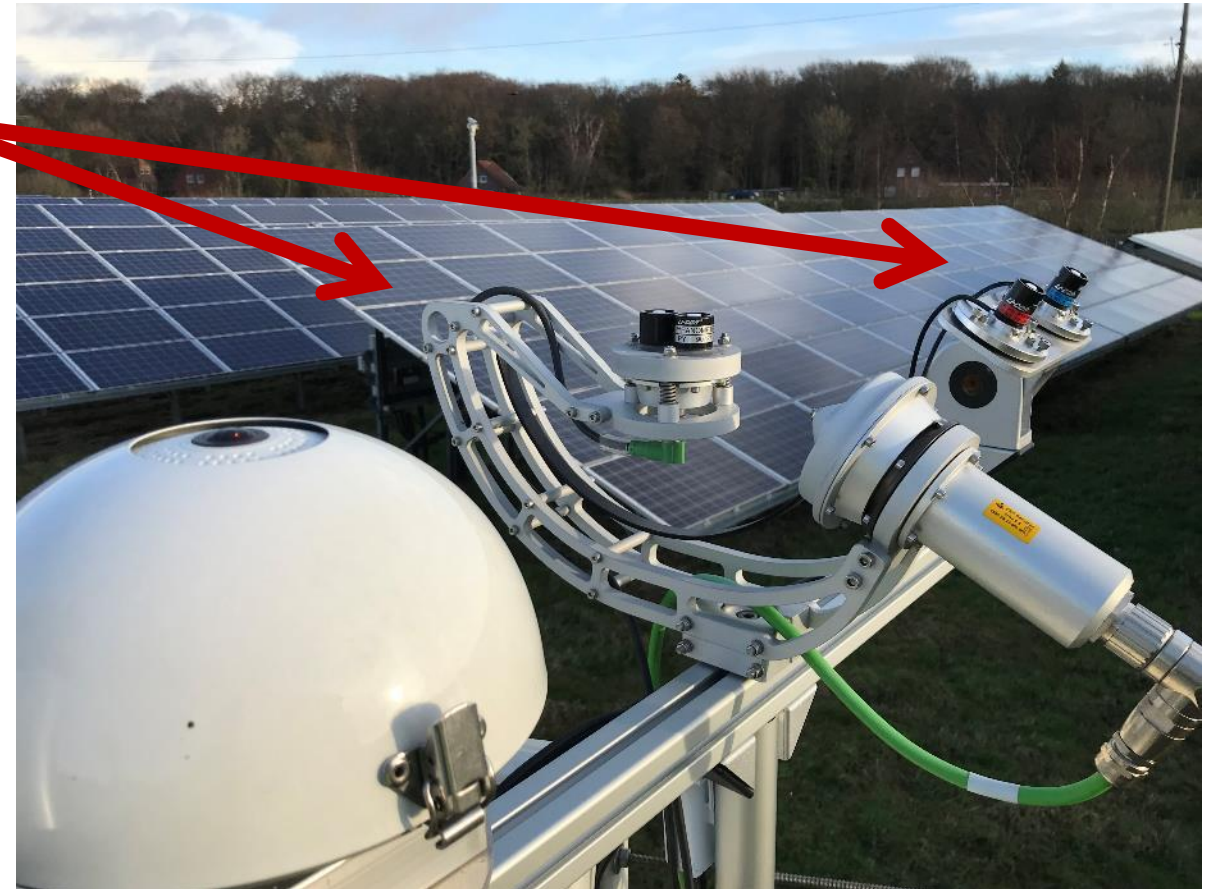
- Solar irradiance sensors (GHI, DHI, DNI, GTI)
- Air temperature and humidity

## All-sky imagers

- Commercial surveillance camera used
- Fish eye lenses with 180° field of view
- Recording images every 30s

## Ceilometers

- 6 atmospheric lidars (ceilometer) measuring cloud height



Photography of Eye2Sky station PVNOR

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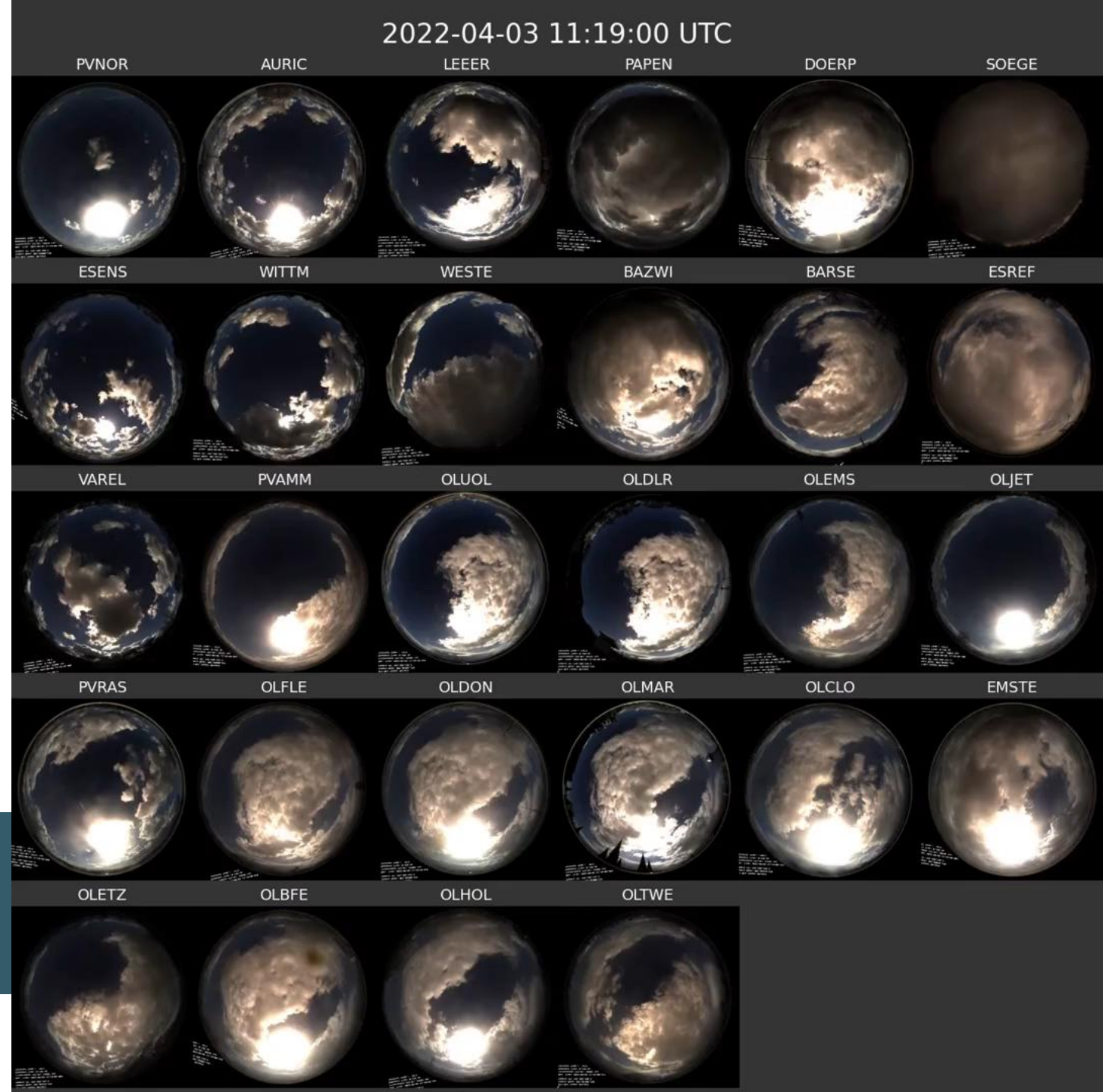
## Ceilometers

- 6 atmospheric lidars (ceilometer) measuring cloud height



Photography of Eye2Sky station PVNOR

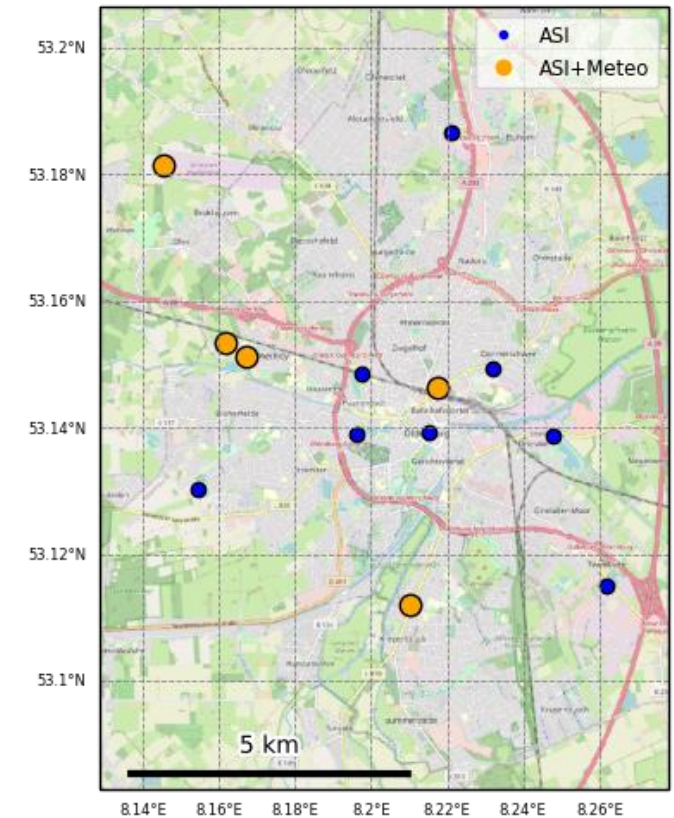
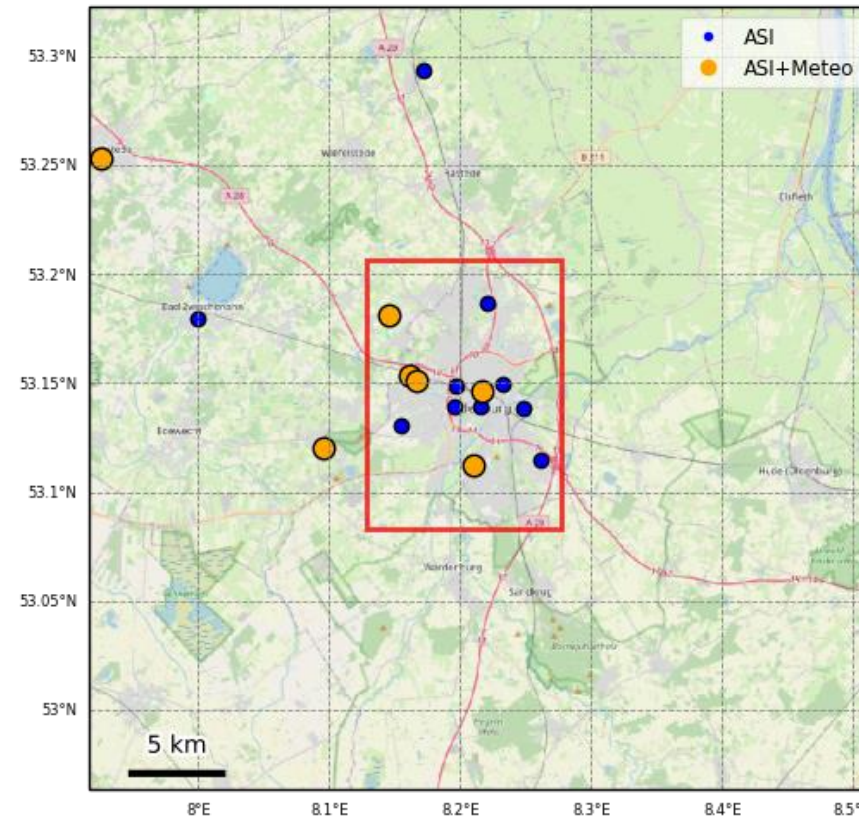
# 2 hours of weather seen by multiple fish eye cameras



## Why cameras?

# Solar irradiance nowcast based on ASI-Network

- Nowcasts for 2022 on 40 x 40 km domain (left)
- 17 ASI used
- Evaluation for city of Oldenburg (10 x 13.75 km, right)
- Grid resolution: 50m



## Nowcasting model for a network of ASI:

- Blum, Niklas (2022): *Nowcasting of Solar Irradiance and Photovoltaic Production Using a Network of All-Sky Imagers*. Dissertation, RWTH Aachen
- Blum, Niklas et al. (2022): *Analyzing Spatial Variations of Cloud Attenuation by a Network of All-Sky Imagers*. Remote Sensing, 14 (22), Seite 5685.

# Comparison of camera, satellite and weather model based data sets



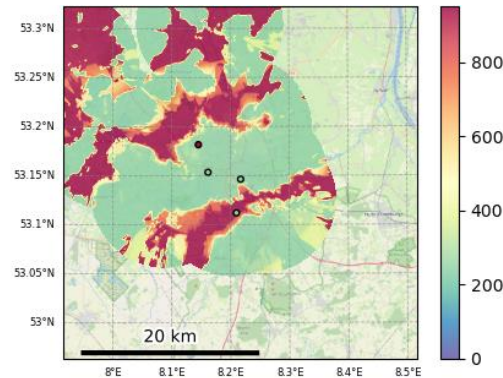
Parameter:  
**GHI**  
Period:  
**06/2022 – 12/2022**

Domain Oldenburg+  
40 km x 40 km

Domain Oldenburg  
13.27 km x 10 km

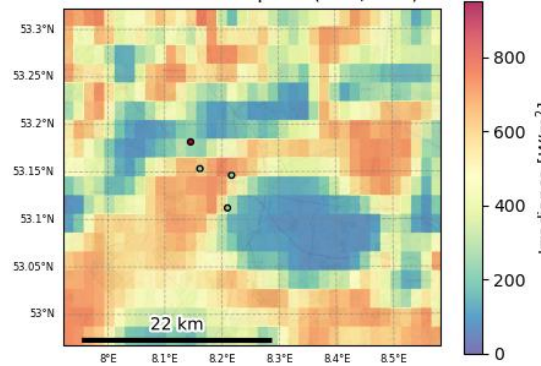
**ASI Network**

$\Delta x = 50\text{m}$   
 $\Delta t = 1\text{ min}$



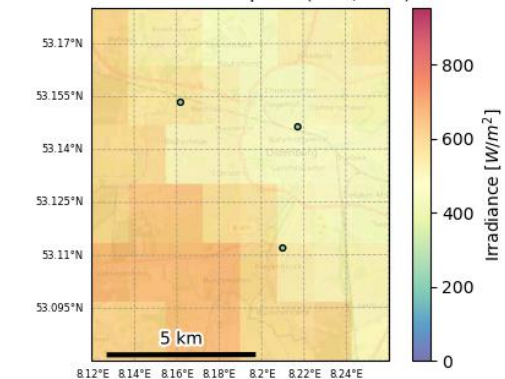
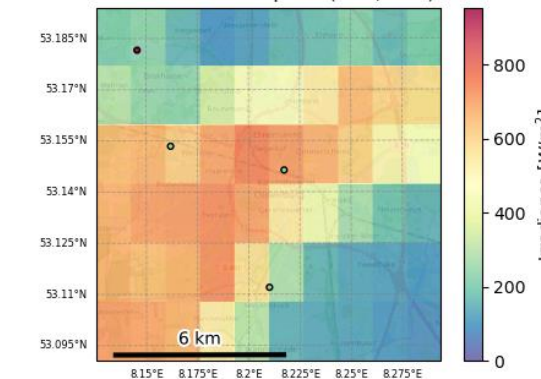
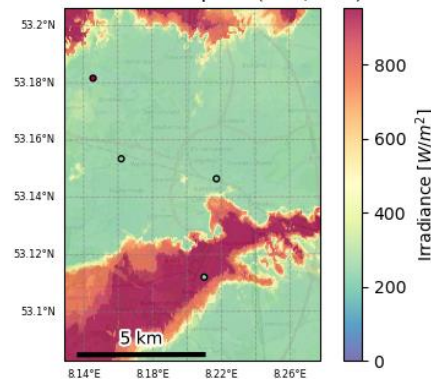
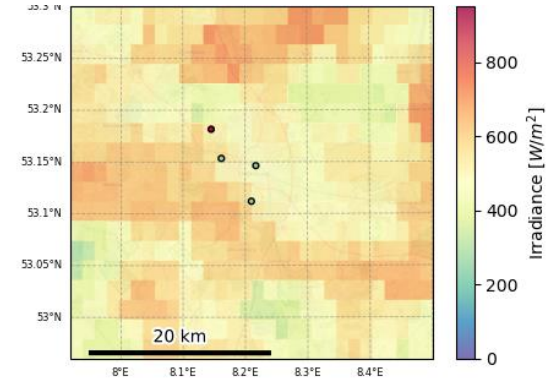
**MSG-SAT**

$\Delta x \sim 2\text{km}$   
 $\Delta t = 15\text{ min}$



**NWP ICON-D2**

$\Delta x \sim 2.2\text{ km}$   
 $\Delta t = 15\text{ min}$   
**Run: 3 UTC**







# Evaluation of spatial variability



- Applying temporal variability indices from literature on solar irradiance maps
- Indices used here:
  - $V = \sigma(\Delta k_c(t))$  (Marquez & Coimbra, 2013)
    - Applied on maps the clear sky index increment is  $\Delta k_c = (k_c(x_i) - k_c(x_i - 1))$  along the x-axis of the map (iterating all rows)
  - $STD = \sigma(k_c)$ 
    - Applied on maps it measures the overall standard deviation of the solar irradiance map

- *Marquez, Ricardo; Coimbra, Carlos F. M. (2013): Proposed Metric for Evaluation of Solar Forecasting Models. In: J. Sol. Energy Eng 135 (1), Artikel 011016. DOI: 10.1115/1.4007496*

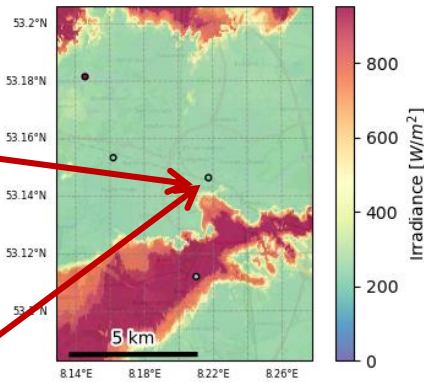
# Spatial Variability Index (broken cloud conditions)



Domain Oldenburg  
13.3 km x 10 km

**ASI Network**  
(Analysis)  
 $\Delta x = 50\text{m}$   
 $\Delta t = 1\text{ min}$  (snapshot)

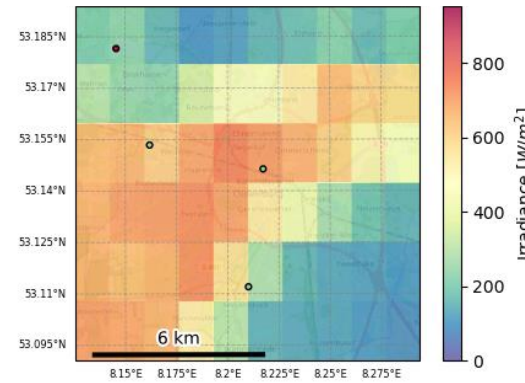
$V = 0.07$   
 $\text{STD} = 0.34$



275 x 200

**MSG-SAT**  
(Analysis, corrected for scan time)  
 $\Delta x \sim 2\text{km}$   
 $\Delta t = 15\text{ min}$  (snapshot)

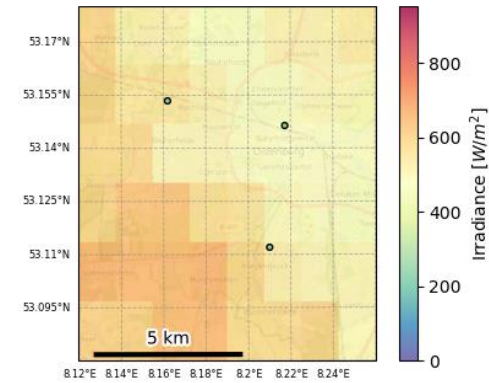
$V = 0.20$   
 $\text{STD} = 0.29$



6 x 10

**NWP ICON-D2**  
(Forecast run 3 UTC)  
 $\Delta x \sim 2.2\text{ km}$   
 $\Delta t = 15\text{ min}$  (average)

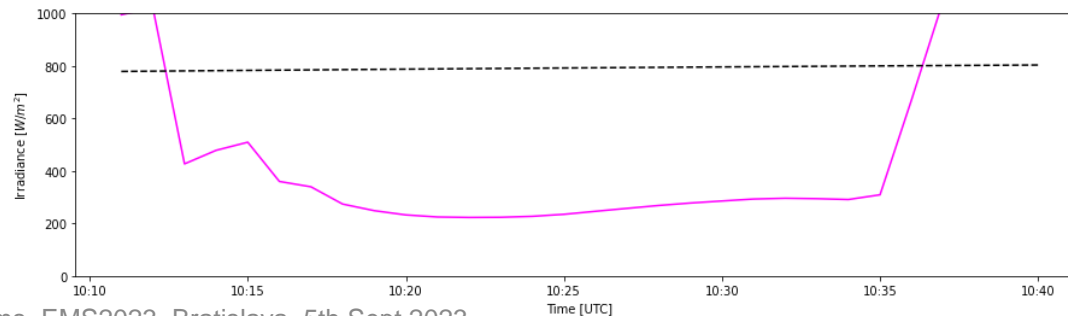
$V = 0.07$   
 $\text{STD} = 0.08$



6 x 8

**In-situ data**  
 $\Delta t = 1\text{ min}$

$V = 0.19$   
 $\text{STD} = 0.25$



Data from 2022-07-27 10:26 UTC

# Spatial Variability Index (small scale cloud conditions)

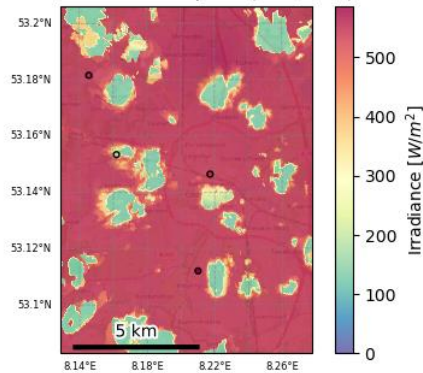


Domain Oldenburg  
13.3 km x 10 km

## ASI Network

$\Delta x = 50\text{m}$   
 $\Delta t = 1\text{ min}$

$V = 0.12$   
 $\text{STD} = 0.28$

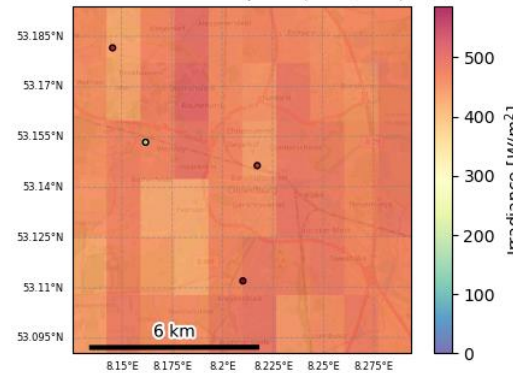


275 x 200

## MSG-SAT

$\Delta x \sim 2\text{km}$   
 $\Delta t = 15\text{ min}$

$V = 0.05$   
 $\text{STD} = 0.04$

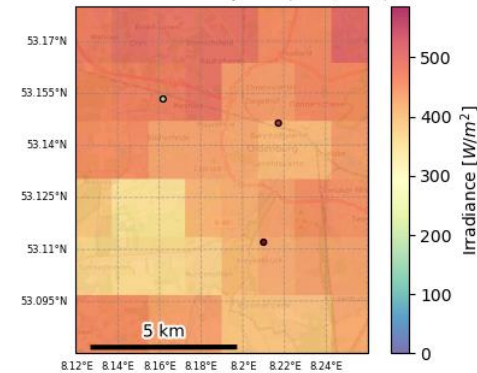


6 x 10

## NWP ICON-D2

$\Delta x \sim 2.2\text{ km}$   
 $\Delta t = 15\text{ min}$   
Run: 3 UTC

$V = 0.05$   
 $\text{STD} = 0.08$

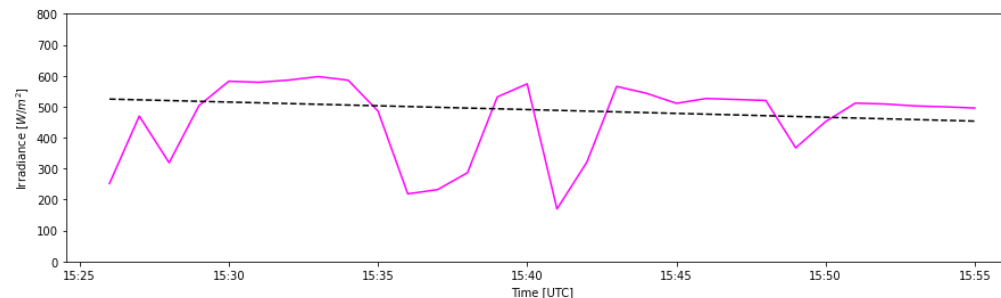


6 x 8

## In-situ data

$\Delta t = 1\text{ min}$

$V = 0.27$   
 $\text{STD} = 0.24$



Data from 2022-07-23 15:41 UTC

# Spatial Variability Index (Overcast conditions)

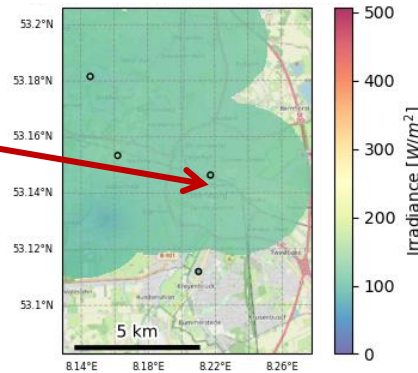


Domain Oldenburg  
13.3 km x 10 km

## ASI Network

$\Delta x = 50\text{m}$   
 $\Delta t = 1\text{ min}$

$V = 0.0$   
 $\text{STD} = 0.02$

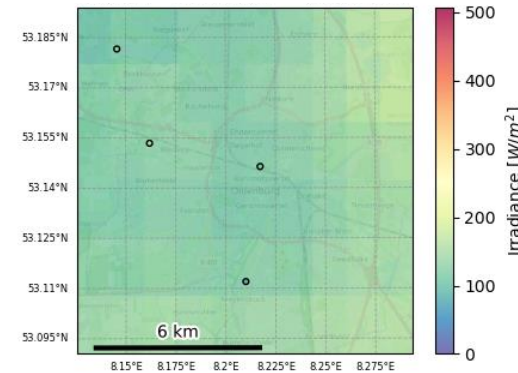


275 x 200

## MSG-SAT

$\Delta x \sim 2\text{km}$   
 $\Delta t = 15\text{ min}$

$V = 0.03$   
 $\text{STD} = 0.04$

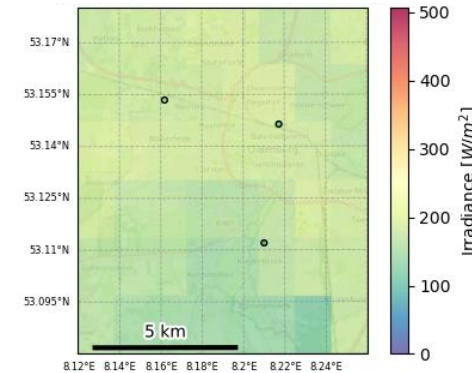


6 x 10

## NWP ICON-D2

$\Delta x \sim 2.2\text{ km}$   
 $\Delta t = 15\text{ min}$   
Run: 3 UTC

$V = 0.03$   
 $\text{STD} = 0.04$

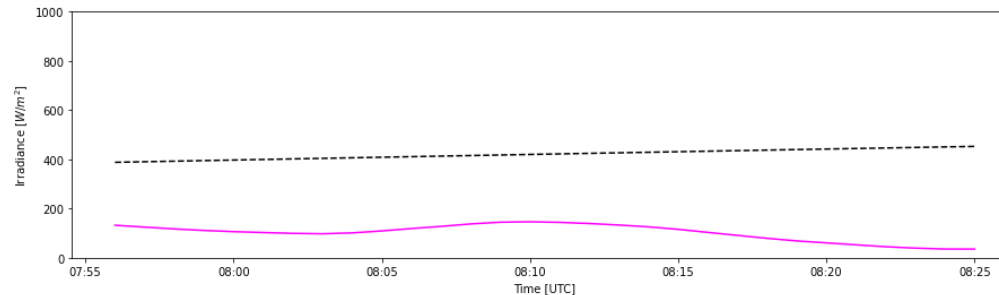


6 x 8

## In-situ data

$\Delta t = 1\text{ min}$

$V = 0.02$   
 $\text{STD} = 0.03$



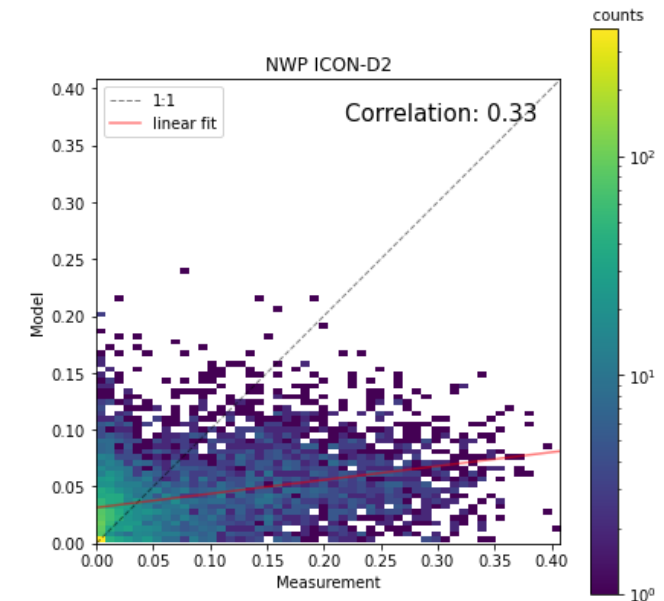
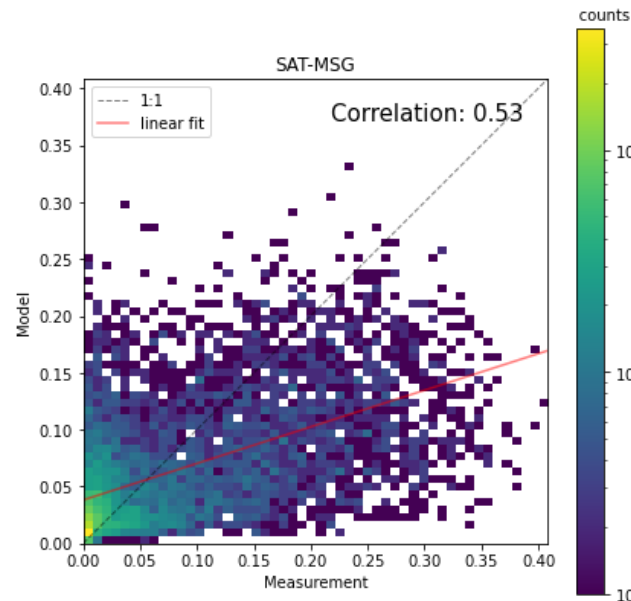
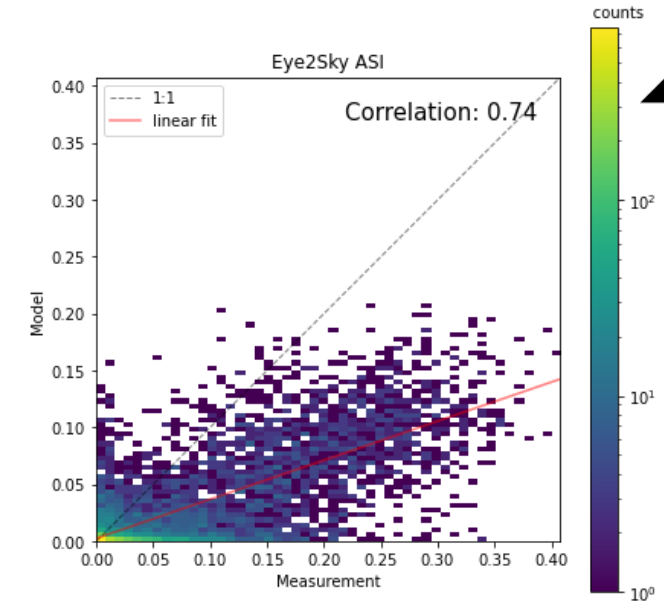
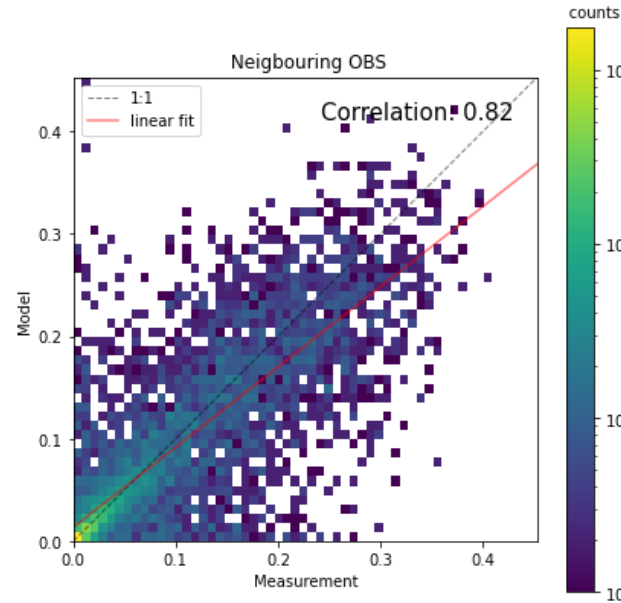
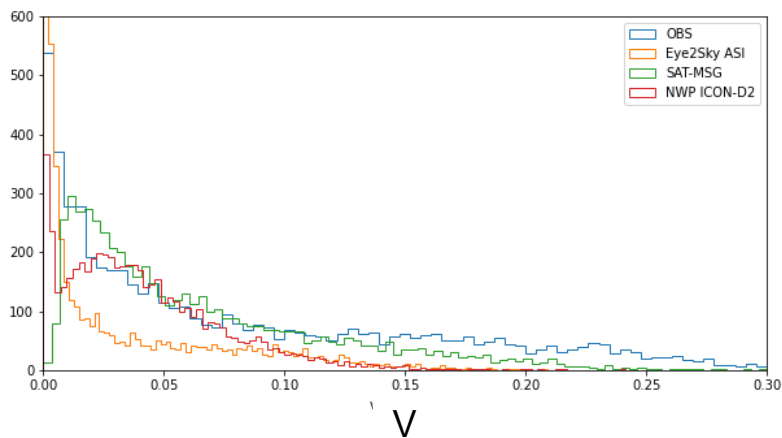
Data from 2022-09-10 08:11 UTC

# Comparison with temporal variability of in-situ measurements

$$V = \sigma(\Delta k_c)$$

- ASI data shows higher correlation than SAT/NWP (*attention: NWP is forecast*)
- The distribution shows better agreement of SAT data with measured timeseries

Histogram of the variability index

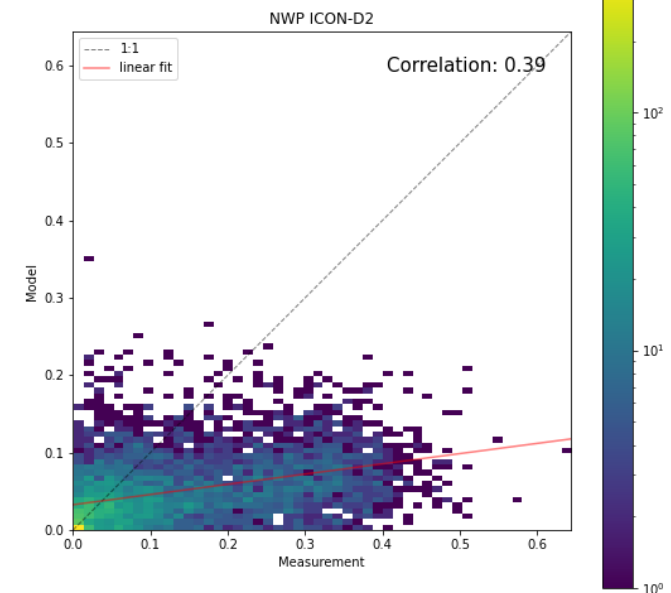
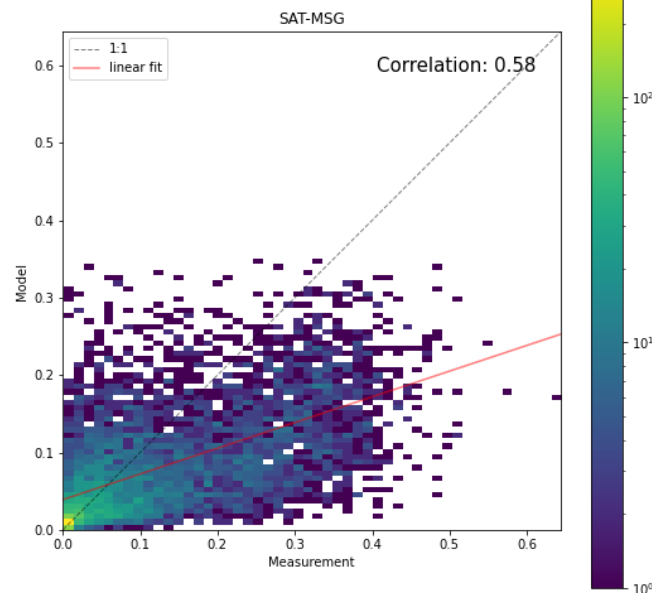
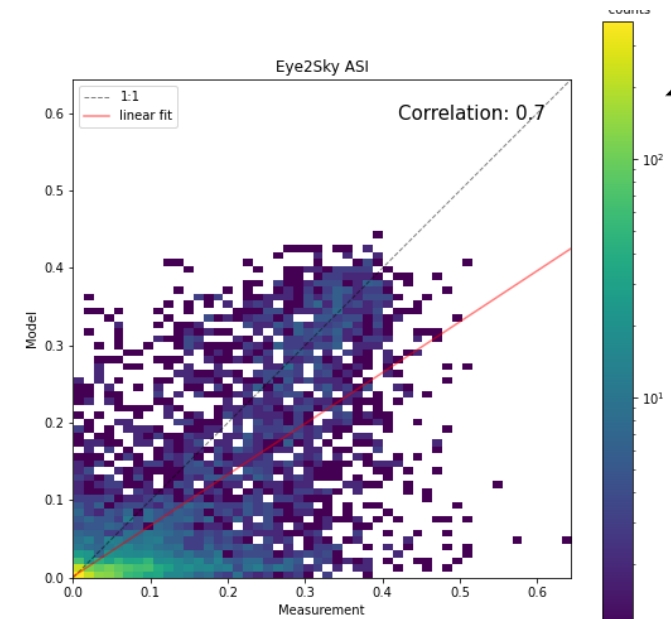
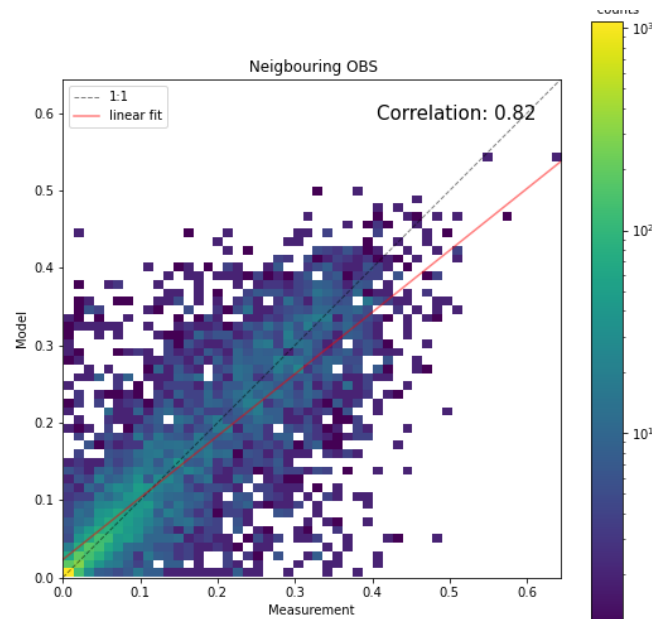
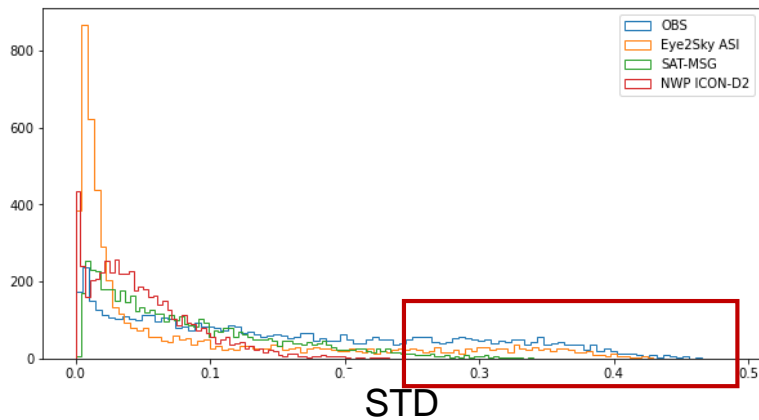


# Comparison with temporal variability of in-situ measurements

$$\text{STD} = \sigma(k_c)$$

- Similar results with even better agreement of ASI with timeseries.
- Distribution shows also better agreement, especially in high variability conditions

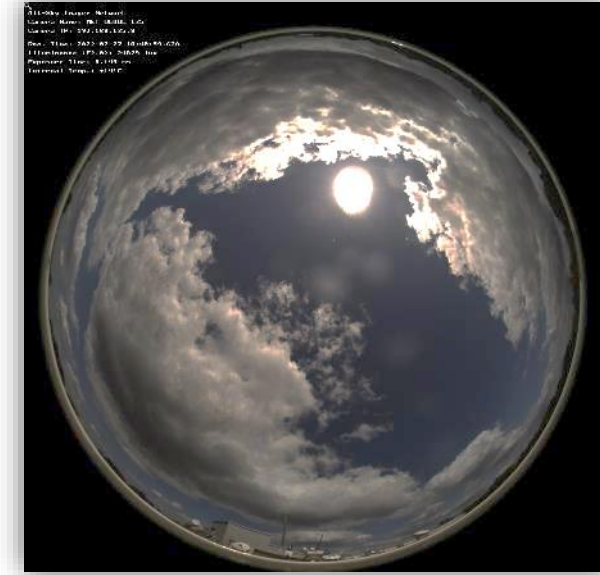
Histogram of the variability index



# Summary

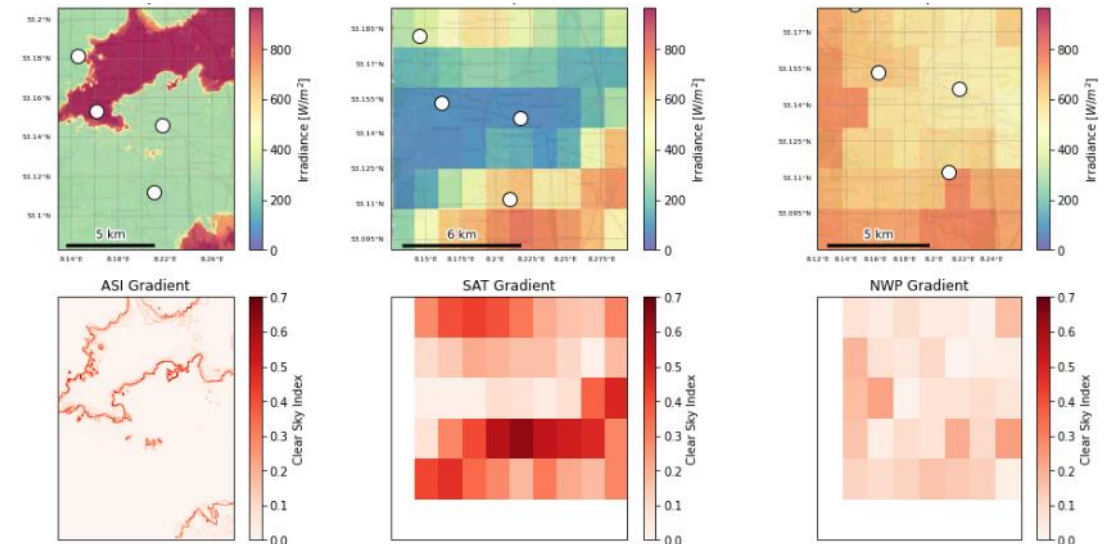
## Conclusions

- ASI-based GHI data provides more details and sharper cloud edges compared to satellite and NWP data
- ASI data is able to resolve small scale cloud patterns
- Simple variability indices used for timeseries show higher correlations with in-situ measurements but cannot show the full potential of the high resolution data sets



## Outlook

- Investigate the relation between temporal and spatial domain, especially with regards to resolution
- Account for cloud motion (speed/direction)
- Apply analysis on forecast data
- Apply analysis on climate model data (EU DestinE project: UseCase Energy Systems)





# Thank you for listening...



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## Website:

<https://www.dlr.de/ve/en/eye2sky>

## Video:

[Portrait of Eye2Sky in 5 Min Video](#)

